

**IN THE CLAIMS:**

*Please find a listing of the claims below. The statuses of the claims are shown in parentheses.*

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1. (Cancelled).

(b) 2. (Currently amended) The method of claim 1, A method for gamut mapping of an input image using a space varying algorithm, comprising:

receiving the input image;

converting the color representations of an image pixel set to produce a corresponding electrical values set;

applying the space varying algorithm to the electrical values set to produce a color-mapped value set;

reconverting the color-mapped value set to an output image; and

wherein the space varying algorithm minimizes a the following variational problem represented by:

$$E(u) = \int_{\Omega} (D^2 + \alpha |\nabla D|^2) d\Omega, \text{ subject to } u \in \mathcal{G}, \text{ wherein } \Omega \text{ is a support of the input}$$

image,  $\mathcal{G}$  is the target gamut,  $\alpha$  is a non-negative real number,  $D = g^*(u - u_0)$ ,  $g$  is a normalized Gaussian kernel with zero mean and a small variance  $\sigma$ ,  $u_0$  is the input image, and  $u$  is the output image.

3. (Original) The method of claim 2, further comprising:

solving the variational problem at a high value of  $\alpha$ ;

solving the variational problem at a low value of  $\alpha$ ; and

averaging the solutions.

4. (Original) The method of claim 3, wherein the step of averaging the solutions comprises using a spatially adaptive weighting scheme, comprising:

$$u_{final}[k,j] = w[k,j]u_{small}[k,j](1-w[k,j])u_{high}[k,j],$$

wherein the weight  $w[k,j]$ , comprises:

$$w[k,j] = \frac{1}{1 + \beta |\nabla g * u_0|^2}, \text{ and}$$

wherein  $\beta$  is a non-negative real number.

5. (Original) The method of claim 2, wherein the variational problem is solved according to:

$$\frac{du}{dt} = \alpha g * \Delta D - g * D, \text{ subject to } u \in \mathcal{B}.$$

6. (Original) The method of claim 2, wherein the space varying algorithm is solved according to:

$$u_y^{n+1} = u_y^n + \tau(\alpha L_y^n - \bar{D}_y^n), \text{ subject to } u_y^n \in \mathcal{B}, \text{ wherein}$$

$$\tau = dt,$$

$$D^n = g * g * (u^n - u_0)$$

$L^n = D_2 * (u^n - u_0)$  and

$D_2 = g_x * g_x + g_y * g_y$

7. (Currently amended) The method of claim 1, A method for gamut mapping of an input image using a space varying algorithm, comprising:  
receiving the input image;  
converting the color representations of an image pixel set to produce a corresponding electrical values set;  
applying the space varying algorithm to the electrical values set to produce a color-mapped value set;  
reconverting the color-mapped value set to an output image; and  
wherein the space varying algorithm minimizes a the following variational problem represented by:

$$E(u) = \int_{\Omega} (\rho_1(D) + \alpha \rho_2(|\nabla D|)) d\Omega, \text{ subject to } u \in \mathcal{G}, \text{ wherein } \rho_1 \text{ and } \rho_2 \text{ are scalar}$$

functions,  $\Omega$  is a support of the image,  $\mathcal{G}$  is the target gamut,  $\alpha$  is a non-negative real number,  $D = g^*(u - u_0)$ ,  $g$  is a normalized Gaussian kernel with zero mean and a small variance  $\sigma$ ,  $u_0$  is the input image, and  $u$  is the output image.

8. (Original) The method of claim 2, further comprising:  
decimating the input image to create one or more resolution layers, wherein the one or more resolution layers comprises an image pyramid; and  
solving the variational problem for each of the one or more resolution layers.

9. (Currently amended) The method of claim [[1]]2, wherein the method is executed in at least one of a camera and a printer.

10. (Currently amended) The method of claim [[1]]2, wherein the method is executed in at least one of a camera and a printer.

11. (Canceled).

12. (Canceled).

13. (Currently amended) The computer readable memory of claim 12, A computer-readable memory for color gamut mapping, comprising an instruction set for executing color gamut mapping steps, the steps, comprising:  
converting first colorimetric values of an original image to second colorimetric values, wherein output values are constrained within a gamut of the output device; using a space varying algorithm that solves an image difference problem; and  
optimizing a solution to the image difference problem, wherein the image difference problem is represented by:

$$E(u) = \int_{\Omega} (D^2 + \alpha |\nabla D|^2) d\Omega$$

subject to  $u \in \mathcal{G}$ , wherein  $\Omega$  is a support of an input image,  $\alpha$  is a non-negative real number,  $\mathcal{G}$  is the target gamut,  $D = g^*(u - u_0)$ ,  $g$  is a normalized Gaussian kernel with zero mean and small variance  $\sigma$ ,  $u_0$  is the input image, and  $u$  is an output image.

14. (Currently amended) The computer-readable memory of claim [[12]]13, wherein the instruction set further comprises steps for:

solving the image difference problem at a high value of  $\alpha$ ;  
solving the image difference problem at a low value of  $\alpha$ ; and  
averaging the solutions.

15. (Original) The computer-readable memory of claim 14, wherein averaging the solutions comprises using a spatially adaptive weighting scheme, comprising:

$$u_{final}[k,j] = w[k,j]u_{small}[k,j](1-w[k,j])u_{high}[k,j],$$

wherein the weight  $w[k,j]$ , comprises:

$$w[k,j] = \frac{1}{1 + \beta |\nabla g * u_0|^2}, \text{ and}$$

wherein  $\beta$  is a non-negative real number.

16. (Currently amended) The computer-readable memory of claim 12, A computer-readable memory for color gamut mapping, comprising an instruction set for executing color gamut mapping steps, the steps, comprising:

converting first colorimetric values of an original image to second colorimetric values, wherein output values are constrained within a gamut of the output device; using a space varying algorithm that solves an image difference problem; and  
optimizing a solution to the image difference problem, wherein the image difference problem is represented by:

$$E(u) = \int_{\Omega} (\rho_1(D) + \alpha \rho_2(|\nabla D|)) d\Omega, \text{ wherein } \rho_1 \text{ and } \rho_2 \text{ are scalar functions.}$$

17. (Currently amended) The computer-readable memory of claim 12; A computer-readable memory for color gamut mapping, comprising an instruction set for executing color gamut mapping steps, the steps, comprising:

converting first colorimetric values of an original image to second colorimetric values, wherein output values are constrained within a gamut of the output device; using a space varying algorithm that solves an image difference problem; and  
optimizing a solution to the image difference problem, wherein the instruction set  
further comprises steps for:

  
decimating the input image to create one or more resolution layers, wherein the one or more resolution layers comprise an image pyramid; and  
solving the image difference problem for each of the one or more resolution layers.

18. (Original) The computer-readable memory of claim 17, wherein the instruction set further comprises steps for:

- (a) initializing a first resolution layer;
- (b) calculating a gradient G for the resolution layer, the gradient G comprising:

$G = \Delta(u - u_0) + \alpha_k(u - u_0)$ , wherein  $\Delta x$  is a convolution of each color plane of x with

$$K_{LAP} = \begin{bmatrix} 0 & 1 & 0 \\ 1 & -2 & 1 \\ 0 & 1 & 0 \end{bmatrix} \text{ and } \alpha_k = \alpha_0 * 2^{2(k-1)}$$

- (c) calculating a normalized steepest descent value  $L_j = L_{j-1} - \mu_0 * \mu_{NSD} * G$ , wherein  $\mu_0$  is a constant;

(d) projecting the value onto constraints  $\text{Proj}_g(L_j)$ , wherein  $\text{Proj}_g(x)$  is a projection of  $x$  into a gamut  $g$ ; and

(e) for a subsequent resolution layer, repeating steps (b) – (d).

19. (Original) A method for image enhancement using gamut mapping, comprising:  
receiving a input image;  
from the input image, constructing an image pyramid having a plurality of resolution layers;

*(b)*  
processing each resolution layer, wherein the processing includes completing a gradient iteration, by:

calculating a gradient  $G$ ;  
completing a gradient descent iteration; and  
projecting the completed gradient descent iteration onto constraints; and  
computing an output image using the processed resolution layers.

20. (Original) The method of claim 19, wherein the gradient  $G$ , comprises:

$$G = \Delta(u-u_0) \alpha_k(u-u_0),$$

wherein  $u$  is the output image,  $u_0$  is the input image, and  $\alpha$  is a non-negative real number.

21. (Original) The method of claim 19, wherein completing the gradient descent iteration comprises calculating:

$$\mu_{NSD} = \frac{\Sigma G^2}{(\Sigma(G * \Delta G) + \alpha_k \Sigma G^2)}; \text{ and}$$

$$L_j = L_{j-1} - \mu_0 \cdot \mu_{NSD} \cdot G,$$

wherein  $\mu_{NSD}$  is a normalized steepest descent parameter,  $\mu_0$  is a constant,  $k$  is a number of resolution layers in the image pyramid, and  $j$  is a specific resolution layer.

22. (Original) The method of claim 19, wherein projecting the completed gradient descent iteration onto the constraints is given by:

$$L_j = \text{Proj}_\mathcal{G}(L_j),$$

wherein  $\text{Proj}_\mathcal{G}(x)$  is a projection of  $x$  into a gamut  $\mathcal{G}$ .

23. (Original) The method of claim 19, wherein constructing the image pyramid, comprises:

smoothing the input image with a Gaussian kernel;  
decimating the input image; and  
setting initial conductive  $L_0 = \max \{S_p\}$ , wherein  $S_p$  is an image with the coarsest resolution layer for the image pyramid.

24. (Original) The method of claim 23, wherein the Gaussian kernel, comprises:

$$K_{PYR} = \begin{bmatrix} 1 & 1 & 1 \\ 16 & 8 & 16 \\ 1 & 1 & 1 \\ 8 & 4 & 8 \\ 1 & 1 & 1 \\ 16 & 8 & 16 \end{bmatrix}$$

25. (Currently amended) The method of claim 19, wherein processing each resolution layer further comprises applying a space varying algorithm to minimize ~~a~~ the following variational problem ~~represented by~~:

$E(u) = \int_{\Omega} (D^2 + \alpha |\nabla D|^2) d\Omega$ , subject to  $u \in \mathcal{S}$ , wherein  $\Omega$  is a support of the

image, ~~g~~ is the target gamut, and  $D = g^*(u - u_0)$ , wherein  $g$  is a normalized Gaussian kernel with zero mean and small variance  $\sigma$ ,  $u_0$  is the input image,  $u$  is the output image, and wherein  $\alpha$  is a non-negative real number.

26. (Original) The method of claim 19, wherein processing each resolution layer comprises applying a space varying algorithm to minimize a variational problem represented by:

$$E(u) = \int_{\Omega} (\rho_1(D) + \alpha \rho_2(|\nabla D|)) d\Omega$$
, subject to  $u \in \mathcal{S}$ , wherein  $\rho_1$  and  $\rho_2$

are scalar function.

27. (Original) The method of claim 26, wherein  $\rho_1$  and  $\rho_2$  are chosen from the group comprising  $\rho(x) = |x|$  and  $\rho(x) = \sqrt{1 + x^2}$ .